Theory/Phenomenology of GPDs

Krešimir Kumerički

University of Zagreb, Croatia



TRANSVERSITY 2017: 5th International Workshop on Transverse Polarization Phenomena in Hard Processes December 11 – 15, 2017 INFN — Frascati National Laboratories, Italy

Fits to harmonics

Global fits results

Neural nets

Conclusion 00000

Outline

Introduction to Generalized Parton Distributions (GPDs) and Deeply Virtual Compton Scattering (DVCS)

Fits to harmonics

Global fits results

Neural nets

Conclusion

Family tree of hadron structure functions



Fits to harmonics

Global fits results

Neural nets 000 Conclusion 00000

Definition of GPDs

• In QCD GPDs are defined as [Müller '92, et al. '94, Ji, Radyushkin '96]

$$F^{q}(x,\eta,t) = \int \frac{dz^{-}}{2\pi} e^{ixP^{+}z^{-}} \langle P_{2}|\bar{q}(-z)\gamma^{+}q(z)|P_{1}\rangle\Big|_{z^{+}=0,z_{\perp}=0}$$
$$\widetilde{F}^{q}(x,\eta,t) = \int \frac{dz^{-}}{2\pi} e^{ixP^{+}z^{-}} \langle P_{2}|\bar{q}(-z)\gamma^{+}\gamma_{5}q(z)|P_{1}\rangle\Big|_{z^{+}=0,z_{\perp}=0}$$

(and similarly for gluons F^g and F^g).



Fits to harmonics

Global fits results 00000000 Neural nets

Conclusion 00000

Some properties of GPDs

• Decomposing into spin-non-flip and spin-flip part:

$$F^{a} = \frac{\bar{u}(P_{2})\gamma^{+}u(P_{1})}{P^{+}}H^{a} + \frac{\bar{u}(P_{2})i\sigma^{+\nu}u(P_{1})\Delta_{\nu}}{2MP^{+}}E^{a} \qquad a = q,g$$
$$\tilde{F}^{a} = \frac{\bar{u}(P_{2})\gamma^{+}\gamma_{5}u(P_{1})}{P^{+}}\tilde{H}^{a} + \frac{\bar{u}(P_{2})\gamma_{5}u(P_{1})\Delta^{+}}{2MP^{+}}\tilde{E}^{a} \qquad a = q,g$$

Fits to harmonics

Global fits results 00000000 Neural nets

Conclusion 00000

Some properties of GPDs

• Decomposing into spin-non-flip and spin-flip part:

$$\begin{split} F^{a} &= \frac{\bar{u}(P_{2})\gamma^{+}u(P_{1})}{P^{+}}H^{a} + \frac{\bar{u}(P_{2})i\sigma^{+\nu}u(P_{1})\Delta_{\nu}}{2MP^{+}}E^{a} \qquad a = q,g\\ \tilde{F}^{a} &= \frac{\bar{u}(P_{2})\gamma^{+}\gamma_{5}u(P_{1})}{P^{+}}\tilde{H}^{a} + \frac{\bar{u}(P_{2})\gamma_{5}u(P_{1})\Delta^{+}}{2MP^{+}}\tilde{E}^{a} \qquad a = q,g \end{split}$$

• "Ji's sum rule" (related to proton spin problem)

$$J^{q} = \frac{1}{2} \int_{-1}^{1} dx \, x \Big[H^{q}(x,\eta,t) + E^{q}(x,\eta,t) \Big]_{t \to 0} \qquad \text{[Ji '96]}$$

Fits to harmonics

Global fits results 00000000 Neural nets

Conclusion 00000

Some properties of GPDs

• Decomposing into spin-non-flip and spin-flip part:

$$\begin{split} F^{a} &= \frac{\bar{u}(P_{2})\gamma^{+}u(P_{1})}{P^{+}}H^{a} + \frac{\bar{u}(P_{2})i\sigma^{+\nu}u(P_{1})\Delta_{\nu}}{2MP^{+}}E^{a} \qquad a = q,g\\ \tilde{F}^{a} &= \frac{\bar{u}(P_{2})\gamma^{+}\gamma_{5}u(P_{1})}{P^{+}}\tilde{H}^{a} + \frac{\bar{u}(P_{2})\gamma_{5}u(P_{1})\Delta^{+}}{2MP^{+}}\tilde{E}^{a} \qquad a = q,g \end{split}$$

• "Ji's sum rule" (related to proton spin problem)

$$J^{q} = \frac{1}{2} \int_{-1}^{1} dx x \Big[H^{q}(x,\eta,t) + E^{q}(x,\eta,t) \Big]_{t \to 0}$$
 [Ji '96]

• Distribution of partons in transversal space

$$ho(x, ec{m{b}}_{\perp}) = \int rac{d^2 ec{\Delta}_{\perp}}{(2\pi)^2} e^{-iec{m{b}}_{\perp} \cdot ec{\Delta}_{\perp}} H(x, 0, t = -ec{\Delta}_{\perp}^2)$$
 [Burkardt '00]

Fits to harmonics

Global fits results 00000000 Neural nets

Conclusion 00000

Access to GPDs via DVCS

- Deeply virtual Compton scattering (DVCS) "gold plated" process of exclusive physics
- DVCS is measured via leptoproduction of a photon



• Interference with Bethe-Heitler process gives unique access to both real and imaginary part of DVCS amplitude.

Fits to harmonics 000000 Global fits results

Neural nets 000 Conclusion 00000

DVCS cross section

$$d\sigma \propto |\mathcal{T}|^2 = |\mathcal{T}_{\mathrm{BH}}|^2 + |\mathcal{T}_{\mathrm{DVCS}}|^2 + \mathcal{I} \; .$$

$$\mathcal{I} \propto \frac{-e_{\ell}}{\mathcal{P}_{1}(\phi)\mathcal{P}_{2}(\phi)} \left\{ c_{0}^{\mathcal{I}} + \sum_{n=1}^{3} \left[c_{n}^{\mathcal{I}} \cos(n\phi) + s_{n}^{\mathcal{I}} \sin(n\phi) \right] \right\},$$
$$\mathcal{T}_{\text{DVCS}}|^{2} \propto \left\{ c_{0}^{\text{DVCS}} + \sum_{n=1}^{2} \left[c_{n}^{\text{DVCS}} \cos(n\phi) + s_{n}^{\text{DVCS}} \sin(n\phi) \right] \right\},$$

 Choosing polarizations (and charges) we focus on particular harmonics:

$$c_{1, ext{unpol.}}^\mathcal{I} \propto \left[F_1 \, \mathfrak{Re} \, \mathcal{H} - rac{t}{4M_p^2} F_2 \, \mathfrak{Re} \, \mathcal{E} + rac{x_ ext{B}}{2-x_ ext{B}} (F_1 + F_2) \, \mathfrak{Re} \, \widetilde{\mathcal{H}}
ight]$$

[Belitsky, Müller et. al '01–'14] • $\mathcal{H}(x_{\mathrm{B}}, t, \mathcal{Q}^2)$, ... — four Compton form factors (CFFs)

Fits to harmonic 000000 Global fits results 00000000 Neural nets 000 Conclusion 00000

Factorization of DVCS \longrightarrow GPDs

• [Collins et al. '98]



• Compton form factor is a convolution:

$${}^{a}\mathcal{H}(x_{\rm B}, t, \mathcal{Q}^{2}) = \int \mathrm{d}x \ C^{a}(x, \frac{x_{\rm B}}{2 - x_{\rm B}}, \frac{\mathcal{Q}^{2}}{\mathcal{Q}_{0}^{2}}) \ H^{a}(x, \frac{x_{\rm B}}{2 - x_{\rm B}}, t, \mathcal{Q}_{0}^{2})$$
$${}^{a=q,G}$$
$$H^{a}(x, \eta, t, \mathcal{Q}_{0}^{2}) - \text{Generalized parton distribution (GPD)}$$

Fits to harmonic 000000 Global fits results

Neural nets 000 Conclusion 00000

Dispersion-relation access to GPDs at LO

[Teryaev '05; K.K., Müller and Passek-K. '07, '08; Diehl and Ivanov '07]

• LO perturbative prediction is "handbag" amplitude

$$\mathcal{H}(\xi, t, \mathcal{Q}^2) \stackrel{\mathrm{LO}}{=} \int_{-1}^{1} dx \; \left(\frac{1}{\xi - x - i\epsilon} - \frac{1}{\xi + x - i\epsilon} \right) H(x, \xi, t, \mathcal{Q}^2)$$

• giving access to GPD on the "cross-over" line $\eta = x$

$$\frac{1}{\pi} \operatorname{\mathfrak{Im}} \mathcal{H}(\xi = x, t, \mathcal{Q}^2) \stackrel{\mathrm{LO}}{=} \mathcal{H}(x, x, t, \mathcal{Q}^2) - \mathcal{H}(-x, x, t, \mathcal{Q}^2)$$

• while dispersion relation connects it to $\mathfrak{Re} \mathcal{H}$

$$\begin{aligned} \mathfrak{Re}\,\mathcal{H}(\xi,t,\mathcal{Q}^2) &= \\ \frac{1}{\pi} \mathrm{PV}\int_0^1 d\xi' \left(\frac{1}{\xi-\xi'}-\frac{1}{\xi+\xi'}\right) \mathfrak{Im}\,\mathcal{H}(\xi',t,\mathcal{Q}^2) + \mathcal{C}_{\mathcal{H}}(t,\mathcal{Q}^2) \end{aligned}$$

Fits to harmonic

Global fits results 00000000 Neural nets

Conclusion 00000

Tomography? (1/3)

- So at LO experimentally accessible observables (CFFs, dσ) are almost* completely determined by H(x, x, t)
- This is good because phenomenology and fits are easier (one variable less)
- This is bad because H(x, x, t) has no probabilistic interpretation, so results cannot be directly used for "tomography"
- To go from cross-over $\eta = x$ GPD section to $\eta = 0$ there are at least three approaches:
- 1. Use NLO precision and evolution to access GPDs away from $\eta = x$ line. (Need large lever arm in Q^2 ; JLab@12GeV?, EIC)

^{*}Up to subtraction constant

Fits to harmonics 000000 Global fits results

Neural nets

Conclusion 00000

Tomography? (2/3)

- 2. Extrapolate $H(x, x, t) \rightarrow H(x, 0, t)$ in a model dependent way:
 - Quark and gluon sea 2D distributions $H(x, \vec{b}_{\perp})$ ([KM] model)



• Sivers effect for valence quarks ([GK] model)



• See also [Dupré, Guidal, Vanderhaeghen '16]

Fits to harmonics

Global fits results

Neural nets

Conclusion 00000

Tomography? (3/3)

- 3. Don't be so focused on probabilistic interpretation.
 - Like, similarly for elastic form factors

$$-rac{1}{6}rac{dG_E(Q^2)}{dQ^2}igg|_{
ightarrow Q^2=0}\longleftrightarrow \langle r^2
angle$$

... extrapolation to Q^2 and interpretation as nucleon 3D mean squared charge radius is also problematic [G. A. Miller '07],

- Anyway, it is form factors F_{1,2}(Q²) that are entering amplitudes of relativistic processes (like DVCS), not radii.
- So let's for a time being concentrate on measuring GPDs as best as we can.

Fits to harmonic

Global fits results 00000000 Neural nets 000 Conclusion 00000

Positivity AND polynomiality

- Probabilistic interpretation is related to the positivity of state vector norm, which is reflected in GPD positivity constraints that are not employed enough in model building
- Polynomiality is easy if GPD is modeled as Radon transform of double distribution (DD)

$$H(x,\eta,t) = x \int d\beta d\alpha \, \delta(x-\beta-\alpha\eta) \, f(\beta,\alpha,t)$$

- OTOH, positivity can be obtained for GPDs represented as overlap of LF wave functions (but gives just DGLAP part x > η)
- So overlap of LFWFs \rightarrow DD via inverse Radon transform \rightarrow gives complete GPD (including ERBL part $x < \eta$). [Hwang and Müller '08, '14], [Chouika et al. '17] [Müller '17]



- Coming soon: COMPASS, JLab12, ... EIC
- See talks by [Van Hulse, Ferrero, Hafidi, Roche, Horn, Joo, ...]

Fits to harmonics 000000 Global fits result 00000000 Neural nets

Conclusion 00000

Experimental coverage (2/2)

Collab	Year	Observables		Kinematics		No. of points		
conub.			x _B	$Q^2 \; [{\rm GeV}^2]$	$ t [\text{GeV}^2]$	total	indep.	
HERMES	2001	$A_{LU}^{\sin \phi}$	0.11	2.6	0.27	1	1	
CLAS	2001	$A_{LU}^{\sin \phi}$	0.19	1.25	0.19	1	1	
CLAS	2006	$A_{UL}^{\sin \phi}$	0.2-0.4	1.82	0.15-0.44	6	3	
HERMES	2006	$A_{C}^{\cos \phi}$	0.08-0.12	2.0-3.7	0.03-0.42	4	4	
Hall A	2006	$\sigma(\phi), \Delta\sigma(\phi)$	0.36	1.5-2.3	0.17-0.33	$4 \times 24 + 12 \times 24$	$_{4\times24+12\times24}$	
CLAS	2007	$A_{LU}(\phi)$	0.11-0.58	1.0-4.8	0.09-1.8	62×12	62×12	
HERMES	2008	$\begin{array}{l} A_{\rm C}^{\cos(0,1)\phi}, \ A_{\rm UT,DVCS}^{\sin(\phi-\phi_S)}, \\ A_{\rm UT,I}^{\sin(\phi-\phi_S)\cos(0,1)\phi}, \\ A_{\rm UT,I}^{\cos(\phi-\phi_S)\sin\phi}, \end{array}$	0.03-0.35	1–10	<0.7	12+12+12 12+12 12	4+4+4 4+4 4	
CLAS	2008	$A_{LU}(\phi)$	0.12-0.48	1.0-2.8	0.1-0.8	66	33	
HERMES	2009	$A_{\mathrm{LU,I}}^{\sin(1,2)\phi}$, $A_{\mathrm{LU,DVCS}}^{\sin\phi}$, $A_{\mathrm{C}}^{\cos(0,1,2,3)\phi}$	0.05-0.24	1.2-5.75	<0.7	18+18+18 18+18+ <i>18</i> +18	6+6+6 6+6+ <i>6</i> +6	
HERMES	2010	$A_{ m UL}^{\sin(1,2,3)\phi},\ A_{ m LL}^{\cos(0,1,2)\phi}$	0.03-0.35	1–10	<0.7	12+12+ <i>12</i> 12+ <i>12</i> +12	4+4+4 4+4+4	
HERMES	2011	$ \begin{array}{l} A_{\mathrm{IT,T}}^{\cos(\phi-\phi_S)\cos(0,1,2)\phi},\\ A_{\mathrm{IT,T}}^{\sin(\phi-\phi_S)\sin(1,2)\phi},\\ A_{\mathrm{IT,T}}^{\cos(\phi-\phi_S)\cos(0,1)\phi},\\ A_{\mathrm{IT,BH+DVCS}}^{\cos(\phi-\phi_S)\cos\phi},\\ A_{\mathrm{IT,BH+DVCS}}^{\sin(\phi-\phi_S)\sin\phi} \end{array} $	0.03–0.35	1–10	<0.7	12+12+12 12+12 12+12 12	4+4+4 4+4 4+4 4	
HERMES	2012	$A_{LU,I}^{\sin(1,2)\phi}$, $A_{LU,DVCS}^{\sin\phi}$, $A_{CO}^{\cos(0,1,2,3)\phi}$	0.03-0.35	1–10	<0.7	18+ <i>18</i> + <i>18</i> 18+18+ <i>18</i> + <i>18</i>	6+ <i>6</i> + <i>6</i> 6+6+ <i>6</i> + <i>6</i>	
CLAS CLAS Hall A	2015 2015 2015	$ \begin{array}{c} \stackrel{C}{A_{LU}}(\phi), A_{UL}(\phi), A_{LL}(\phi) \\ \sigma(\phi), \Delta\sigma(\phi) \\ \sigma(\phi), \Delta\sigma(\phi) \end{array} $	0.17-0.47 0.1-0.58 0.33-0.40	1.3–3.5 1–4.6 1.5–2.6	0.1–1.4 0.09–0.52 0.17–0.37	166+166+166 2640+2640 480+600	$\substack{166+166+166\\2640+2640\\240+360}$	

Fits to harmonic 000000 Global fits results 00000000 Neural nets 000 Conclusion 00000

Alternative processes for GPD access

• Deeply virtual meson production (DVMP) $\gamma^* p \rightarrow Mp$.



- Theory more "dirty" than for DVCS (second "soft" function appears: meson distribution amplitude)
- Different mesons enable access to different flavours of GPDs [Goloskokov, Kroll]
- First global DVCS+DVMP NLO fit [Lautenschlager, Müller, Schäfer '13]
- double DVCS $\gamma^* p \rightarrow \gamma^* p$, timelike DVCS, ... see talk by [T. Horn]

 Ds and DVCS
 Fits to harmonics
 Global fits results
 Neural nets

 ••••••••
 •••••••
 ••••••
 •••••

 Case #1: CLAS cross-section (\$\phi\$-space)



- KM15 model (global refit including this data): $\chi^2/\text{npts} = 1032.0/1014$ for $d\sigma$ and 936.1/1012 for $\Delta\sigma$
- looks excellent

Fits to harmonics 00000 Global fits results 00000000 Neural nets 000 Conclusion 00000

Going from ϕ - to *n*-space

• It is often convenient to work with weighted harmonics

$$\sigma^{\sin n\phi, w} \equiv rac{1}{\pi} \int_{-\pi}^{\pi} dw \, \sin(n\phi) \, \sigma(\phi) \; ,$$

with specialy weighted Fourier integral measure

$$egin{aligned} & \mathsf{d} \mathsf{w} \equiv rac{2\pi \mathcal{P}_1(\phi)\mathcal{P}_2(\phi)}{\int_{-\pi}^{\pi} \mathsf{d} \phi \, \mathcal{P}_1(\phi)\mathcal{P}_2(\phi)} \mathsf{d} \phi \ , \end{aligned}$$

thus cancelling strongly oscillating factors $1/(\mathcal{P}_1(\phi)\mathcal{P}_2(\phi))$ in Bethe-Heitler and interference terms in $d\sigma$.

• Highest extractable harmonics:

	CLAS			Hall A		
	sine	cosine		sine	cosine	
$\Delta \sigma^w$	1.6 ± 0.8	0.8 ± 1.1		1.4 ± 0.5	$\textbf{0.3}\pm\textbf{0.8}$	
$d\sigma^w$	$\textbf{0.7}\pm\textbf{0.9}$	1.6 ± 1.2		1.2 ± 1.1	2.1 ± 0.8	

Fits to harmonics

Global fits result: 00000000 Neural nets

Conclusion 00000

Case #1: CLAS cross-section (n-space)





• $\chi^2/\text{npts} = \frac{62.2}{48}$ for $d\sigma^{\cos\phi,w}$

(O.K. but not so perfect as in ϕ -space)

Fits to harmonics

Global fits result: 00000000 Neural nets

Conclusion 00000

Case #1: CLAS cross-section (n-space)





• $\chi^2/\text{npts} = \frac{62.2}{48}$ for $d\sigma^{\cos\phi,w}$

(O.K. but not so perfect as in ϕ -space)

In addition to $\chi^2/{\rm d.o.f.}$ use:

$$\mathsf{pull} \equiv \frac{1}{\sqrt{N}} \sum_{i=1}^{N} \frac{f(x_i) - y_i}{\Delta y_i}$$

Conclusion 00000

Propagation of uncertainties to harmonics

- Consider three types of uncertainty:
 - 1. uncorrelated point-to-point uncertainty (absolute size ϵ)
 - 2. correlated normalization uncertainty (relative size ϵ)
 - 3. correlated modulated (ϕ -dependent) uncertainty (e.g., relative size $\epsilon \cos(\phi)$)
- Uncorrelated uncertainty: $\Delta c_k = \sqrt{2/N} \ \epsilon$
- Normalization uncertainty: $\Delta c_k/c_k = \epsilon$
- Correlated modulated uncertainty: more complicated, but for hierarchical case $c_0 \gg c_1 \gg \cdots$ one obtains

$$\frac{\Delta c_0}{c_0} = \frac{c_1}{2c_0}\epsilon, \qquad \frac{\Delta c_1}{c_1} = \frac{c_0}{c_1}\epsilon$$

i.e. we have enhancement of uncertainty for subleading harmonics!

$$(c_0+c_1\cos\phi+\cdots)\times(1+\epsilon\cos\phi)=c_0\left(1+rac{c_1}{2c_0}\epsilon\right)+c_1\left(1+rac{c_0}{c_1}\epsilon\right)\cos\phi$$

Fits to harmonics

Global fits results 00000000 Neural nets 000 Conclusion 00000

Modulated correlated error in the wild

Hall A [M. Defurne et. al 2015] discussing systematic uncertainties:

tematic error from the parameter choice to be 1%.

FIG. 20. Difference in % between the cross section extracted with the squared DVCS amplitude term and with the Re $C^{T,V}$ term for $x_B = 0.37$, $Q^2 = 2.36$ GeV² and -t = 0.33 GeV². The ϕ profile of the difference is a consequence of the small $\cos\phi$ and $\cos2\phi$ dependences of the Re $C^{T,V}$ kinematic coefficient. Both extractions give almost the same reduced $\chi^2/dof=0.94$ (nominal) and 0.93 (alternate) for the entire Kin2 setting.

Systematic uncertainty	Value 8	Section
HRS acceptance cut	1%	IV A
Electron ID	0.5%	IVD
HRS multitrack	0.5%	IVD
Multi-cluster	0.4%	IVD
Corrected luminosity	1%	IVD
Fit parameters	1%	VIB
Radiative corrections	2%	V
Beam polarization	2%	III A 3
Total (helicity-independent)	2.8%	
Total (helicity-dependent)	3.4%	

TABLE V. Normalization systematic uncertainties in the extracted photon electroproduction cross sections. The systematic error coming from the fit parameter choice is not a normalization error per se, but we consider that 1% is an upper limit for this error on all kinematic bins. The helicity-dependent cross sections have an extra uncertainty stemming from the beam polarization measurement. The last column gives the section in which each systematic effect is discussed.

Fits to harmonics

Global fits result: 00000000 Neural nets 000 Conclusion 00000

Hall A $cos(\phi)$ harmonics

(Syst added *linearly* on top of stat.)

Fits to harmonic 000000

Global fits results

Neural nets 000 Conclusion 00000

Hybrid GPD models for global fits

- Sea quarks and gluons modelled using SO(3) partial wave expansion in conformal GPD moment space + Q^2 evolution.
- Valence quarks model CFFs directly (ignoring Q^2 evolution):

$$\Im \mathfrak{M} \mathcal{H}(\xi, t) = \pi \left[\frac{4}{9} H^{u_{\text{val}}}(\xi, \xi, t) + \frac{1}{9} H^{d_{\text{val}}}(\xi, \xi, t) + \frac{2}{9} H^{\text{sea}}(\xi, \xi, t) \right]$$
$$H(x, x, t) = n r 2^{\alpha} \left(\frac{2x}{1+x} \right)^{-\alpha(t)} \left(\frac{1-x}{1+x} \right)^{b} \frac{1}{\left(1 - \frac{1-x}{1+x} \frac{t}{M^{2}} \right)^{p}}.$$

- $\mathfrak{Re}\,\mathcal{H}$ determined by dispersion relations
- 15 free parameters in total for H, \tilde{H} , E, \tilde{E} .

GPDs and DVCS 00000000	Fits to harmonics 000000		Global fits results ○●○○○○○○			Neural nets 000		
Model	KM09a	KM09b	KM10	KM10a	KM10b	KMS11	KMM12	KM15
free params.	{3}+(3)+5	{3}+(3)+6	{3}+15	{3}+10	$\{3\}+15$	NNet	$\{3\}+15$	$\{3\}+15$
$\chi^2/d.o.f.$	32.0/31	33.4/34	135.7/160	129.2/149	115.5/126	13.8/36	123.5/80	240./275
F ₂	{85}	{85}	{85}	{85}	{85}		{85}	{85}
$\sigma_{\rm DVCS}$	(45)	(45)	51	51	45		11	11
$d\sigma_{ m DVCS}/dt$	(56)	(56)	56	56	56		24	24
$A_{LU}^{\sin \phi}$	12 + 12	12 + 12	12	16	12 + 12		4	13
$A_{LU,I}^{\sin\phi}$			18	18		18	6	6
$A_C^{\cos 0\phi}$							6	6
$A_C^{\cos \phi}$	12	12	18	18	12	18	6	6
$\Delta \sigma^{\sin \phi, w}$			12				12	63
$\sigma^{\cos 0\phi, w}$			4				4	58
$\sigma^{\cos\phi,w}$			4				4	58
$\sigma^{\cos\phi,w}/\sigma^{\cos0\phi,w}$		4			4			
$A_{UL}^{\sin \phi}$							10	17
$A_{LL}^{\cos 0\phi}$							4	14
$A_{LL}^{\cos \phi}$								10
$A_{UT,I}^{\sin(\phi-\phi_S)\cos\phi}$							4	4

- [K.K., Müller, et al. '09–'15]
- Other approach: recursive fit [Goldstein, Gonzalez, Liuti '11]

Fits to harmonic 000000 Global fits results

Neural nets 000 Conclusion 00000

KM models are available at WWW

 http://calculon.phy.hr/gpd/ — binary code for cross sections

% xs.exe

xs.exe ModelID Charge Polarization Ee Ep xB Q2 t phi

returns cross section (in nb) for scattering of lepton of energy Ee on unpolarized proton of energy Ep. Charge=-1 is for electron.

ModelID is one of 1 KM09a - arXiv:0904.0458 fit without Hall A data, 2 KM09b - arXiv:0904.0458 fit with Hall A harmonics ratio, 3 KM10 - arXiv:1105.0899 fit with Hall A harmonics 4 KM10a - arXiv:1105.0899 fit without Hall A data 5 KM10b - arXiv:1105.0899 fit with Hall A harmonics ratio 6 KMM12 - arXiv:1301.1230 fit with Hall A harmonics and polarized target 7 KM15 - arXiv:1512.09014 fit now includes 2015 CLAS and Hall A data

xB Q2 t phi -- usual kinematics (phi is in Trento convention)

% xs.exe 1 -1 1 27.6 0.938 0.111 3. -0.3 0

0.18584386497251

• Plots of all CFFs available; numerical values soon to come

Fits to harmonic 000000 Global fits results

Neural nets

Conclusion 00000

2006 vs 2015 Hall A cross-sections

• Improvement of global $\chi^2/d.o.f.$ 123.5/80 \rightarrow 240./275

Fits to harmonic 000000 Global fits results

Neural nets 000 Conclusion 00000

2015 vs 2017 Hall A cross-sections (1/2)

• see talk by [J. Roche]

 Predictions of KM15 model fail for low-Q² 2017 Hall A measurements!
 PDs and DVCS
 Fits to harmonics
 Global fits results
 Neural nets

 2015 vs 2017 Hall A cross-sections (2/2)

- New data are a challenge for global fit
- ... but have gluons really been glimpsed by Hall A?

Fits to harmonic 000000 Global fits results

Neural nets

Conclusion 00000

$\chi^2/npts$ and pull values

Collaboration	Observable	n .	KMM1	2	KM15		
Collaboration	Observable	"pts	$\chi^2/n_{ m pts}$	pull	$\chi^2/n_{ m pts}$	pull	
ZEUS	$\sigma_{\rm DVCS}$	11	0.49	-1.76	0.51	-1.74	
ZEUS,H1	$d\sigma_{ m DVCS}/dt$	24	0.97	0.85	1.04	1.37	
HERMES	$A_{\rm C}^{\cos 0\phi}$	6	1.31	0.49	1.24	0.29	
HERMES	$A_{C}^{\cos \phi}$	6	0.24	-0.56	0.07	-0.20	
HERMES	$A_{LU,I}^{\sin \phi}$	6	2.08	-2.52	1.34	-1.28	
CLAS	$A_{ m LU}^{\sin \phi}$	4	1.28	2.09			
CLAS	$A_{LU}^{\sin \phi}$	13			1.24	0.63	
CLAS	$\Delta \sigma^{\sin \phi, w}$	48			0.41	-1.66	
CLAS	$d\sigma^{\cos 0\phi,w}$	48			0.16	-0.21	
CLAS	$d\sigma^{\cos\phi,w}$	48			1.16	6.36	
Hall A	$\Delta \sigma^{\sin \phi, w}$	12	1.06	-2.55			
Hall A	$d\sigma^{\cos 0\phi,w}$	4	1.21	2.14			
Hall A	$d\sigma^{\cos\phi,w}$	4	3.49	-0.26			
Hall A	$\Delta \sigma^{\sin \phi, w}$	15			0.81	-2.84	
Hall A	$d\sigma^{\cos 0\phi,w}$	10			0.40	0.92	
Hall A	$d\sigma^{\cos\phi,w}$	10			2.52	-2.42	
HERMES, CLAS	$A_{III}^{\sin \phi}$	10	1.90	-1.89	1.10	-1.94	
HERMES	$A_{LL}^{\cos 0\phi}$	4	3.44	2.17	3.19	1.99	
HERMES	$A_{UT,I}^{\sin(\phi-\phi_S)\cos\phi}$	4	0.90	0.61	0.90	0.71	
CLAS	$A_{\rm UL}^{\sin \phi}$	10			0.76	0.38	
CLAS	$A_{LL}^{\cos 0\phi}$	10			0.50	-0.22	
CLAS	$A_{LL}^{\overline{\cos}\phi}$	10			1.54	2.40	

Fits to harmonic 000000

Global fits results 00000000 Neural nets

Conclusion 00000

Problems with standard fitting approaches

- 1. Choice of fitting function introduces theoretical bias leading to systematic error which cannot be estimated (and is surely much larger for $GPDs(x, \eta, t)$ than for PDFs(x).
- 2. Propagation of uncertainties from experiment to fitted function is difficult. Errors in actual experiments are not always Gaussian.

Fits to harmonic 000000

Global fits results 00000000 Neural nets

Conclusion 00000

Problems with standard fitting approaches

- 1. Choice of fitting function introduces theoretical bias leading to systematic error which cannot be estimated (and is surely much larger for $GPDs(x, \eta, t)$ than for $PDFs(x) \rightarrow NNets$
- Propagation of uncertainties from experiment to fitted function is difficult. Errors in actual experiments are not always Gaussian.→ Monte Carlo error propagation

• Essentially a least-square fit of a complicated many-parameter function. $f(x) = \tanh(\sum w_i \tanh(\sum w_j \cdots)) \Rightarrow$ no theory bias

Fits to harmonic 000000 Global fits results 00000000 Neural nets

Conclusion 00000

Resulting neural network CFFs

Fits to harmonics 000000 Global fits result

Neural nets

Conclusion •0000

CFFs from various fits

(from [K.K., Moutarde and Liuti '16])

Fits to harmonics 000000 Global fits results 00000000 Neural nets 000 Conclusion

CFFs from various fits (II)

Comparing $\mathfrak{Im} \mathcal{H}$ and D-term for a typical JLab kinematics:

Fits to harmonics 000000 Global fits results 00000000 Neural nets

Conclusion

CFFs from various fits (II)

Comparing $\mathfrak{Im} \mathcal{H}$ and D-term for a typical JLab kinematics:

• We need access to ERBL region. (\rightarrow double DVCS?)

Fits to harmonic 000000

Global fits results 00000000 Neural nets 000 Conclusion

Going beyond first approximations

Published DVCS data (apart from new 2017 Hall A) is well described within leading order, leading twist and other approximations. Some available corrections:

- NLO evolution and NLO coefficient functions
- finite-t and target mass corrections [Braun et al. '14]
- twist-3 GPDs
- massive charm [Noritzsch '03]
- small-x resummation

can maybe be absorbed in redefinition of GPDs (think "DVCS scheme" factorization), but have to be taken into account when working with more processes (striving towards universal GPDs) and with more precise future data.

Fits to harmonic

Global fits results

Neural nets 000 Conclusion

DVCS at EIC

- Future polarized electron-ion collider (EIC) will provide unique insight into sea GPDs.
- [Aschenauer, Fazio,

K.K., Müller '13] fit to simulated DVCS data at 20 GeV \times 250 GeV

- Improved knowledge of low-t quark and gluon GPDs H
- Improved knowledge of sea quark GPD E

- Global fits of all proton DVCS data using flexible hybrid models were in healthy shape until 2017 (some tension for first $cos(\phi)$ harmonic of JLab cross sections).
- New Hall A 2017 data present a challenge
- Standard global model fitting and neural networks approach are complementing each other

- Global fits of all proton DVCS data using flexible hybrid models were in healthy shape until 2017 (some tension for first $cos(\phi)$ harmonic of JLab cross sections).
- New Hall A 2017 data present a challenge
- Standard global model fitting and neural networks approach are complementing each other

The End